

# MrPaw

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# Are US non-voters becoming more Republican?

## Blue Rose research says yes:

“Politically disengaged voters have become much more Republican, And because less-engaged voters swung away from [Democrats], an expanded electorate meant a more Republican electorate.”

[Blue Rose Research, 2024] (On Ezra Klein show, major professional pollsters)

Several factors drive the disagreement:

- The problem is very hard (it’s difficult to poll non-voters)
- Different data sources
  - Blue Rose aggregates its own private data
  - The *On Data and Democracy* posts use public data, e.g. the cooperative election study (CES).
- **Very different statistical methods:** ★
  - Blue Rose uses Bayesian hierarchical modeling
  - The CES uses calibration weighting

**Our work won’t resolve the dispute.** (Anyway, we’d need access to Blue Rose’s private data and modeling to even try.)

But we can form a like-to-like comparison between the methodologies. (And hope that Blue Rose tries our software package.)

## *On Data and Democracy* says no:

“Claims of a decisive pro-Republican shift among the overall non-voting population are not supported by the most reliable, large-scale post-election data currently available.”

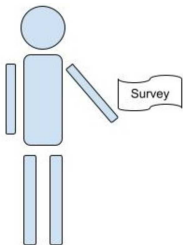
[Bonica et al., 2025] (Berkeley professor co-author, major professional researchers)

# The basic problem

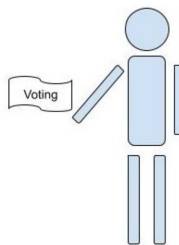
We have a survey population, for whom we observe:

- Covariates  $x$  (e.g. race, gender, zip code, age, education level)
- Responses  $y$  (e.g. A binary response to “do you support policy such-and-such”)

We want the average response in a target population, in which we observe only covariates.



Observe  $(x_s, y_s)$  for  $s = 1, \dots, \mathcal{N}_S$



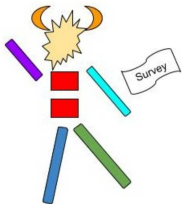
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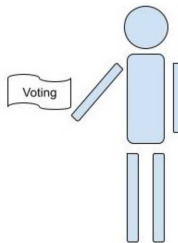
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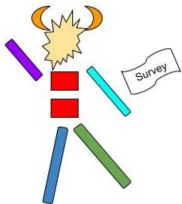
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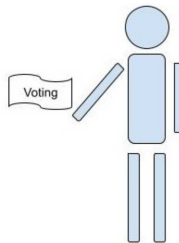
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**The problem is that the populations are very different.**

Our survey results may be biased.

How can we use the covariates to say something about the target responses?

## Two approaches

We want  $\mu := \frac{1}{N_T} \sum_{n \in \mathcal{N}_T} y_p$ , but don't observe  $y_p$  in the target population.

- Assume  $p(y|x)$  is the same in both populations,
- But the distribution of  $x$  may be different in the survey and target.

### Calibration weighting

Choose “calibration weights”  $w_s$   
(e.g. raking weights)

$$\hat{\mu}_{\text{CAL}} = \frac{1}{N_S} \sum_{n \in \mathcal{N}_S} w_s y_s$$

Dependence on  $y_s$  is obvious  
( $w_s$  typically chosen using only  $x$ )

Weights give interpretable diagnostics:

- Frequentist variability
- Partial pooling
- Regressor balance

### Bayesian hierarchical modeling (MrP)

Choose a model  $\mathcal{P}(y|x, \theta)$  and prior  $\mathcal{P}(\theta)$   
(e.g. Hierarchical logistic regression)

Take  $\hat{y}_p = \mathbb{E}_{\mathcal{P}(\theta | \text{Survey data})} [y|x_p]$  and  
$$\hat{\mu}_{\text{MRP}} = \frac{1}{N_T} \sum_{n \in \mathcal{N}_T} \hat{y}_p$$

Dependence on  $y_s$  very complicated  
(Typically via MCMC draws from  
 $\mathcal{P}(\theta | \text{Survey data})$ )

**Black box**

We open the MrP black box, and provide versions of all these diagnostics, for nonlinear hierarchical models fit with MCMC.

- Blue Rose Research. 2024 Election Retrospective Presentation.  
<https://data.blueroseresearch.org/2024retro-download>, 2024. Accessed on 2024-10-26.
- A. Bonica, R. Fordham, J. Grumbach, and E. Tiburcio. Did non-voters really flip Republican in 2024? The evidence says no.  
<https://data4democracy.substack.com/p/did-non-voters-really-flip-republican>, April 2025.